

# STATISTICAL DOWNSCALING: *IS PAST PERFORMANCE AN INDICATOR OF FUTURE RESULTS?*

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1



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2



TEXAS TECH

3



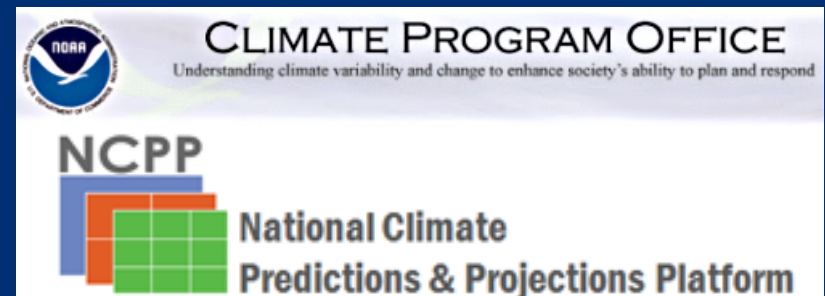
Univ. of OKLAHOMA

4 Princeton Univ.

5 DRC, inc.

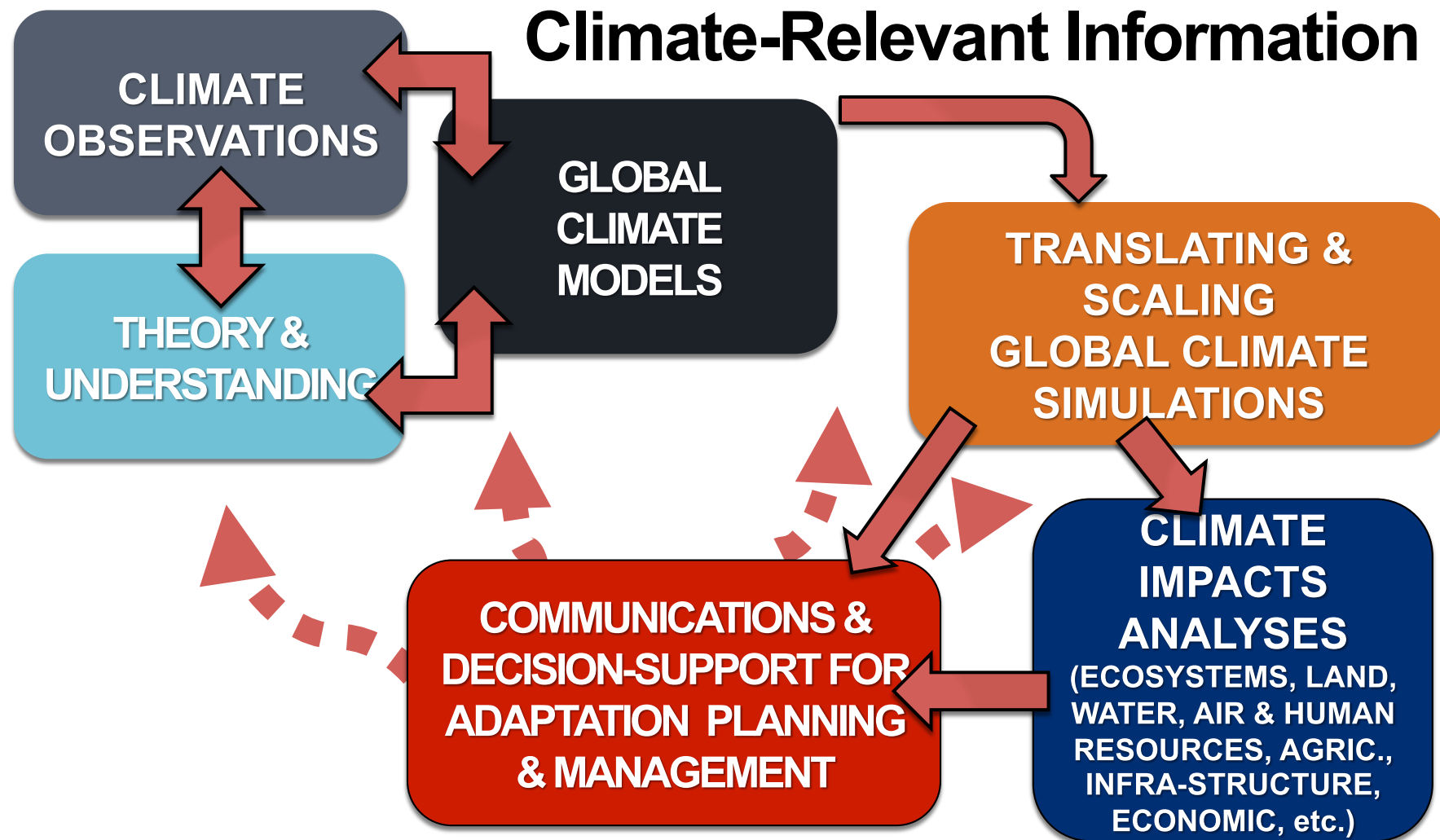


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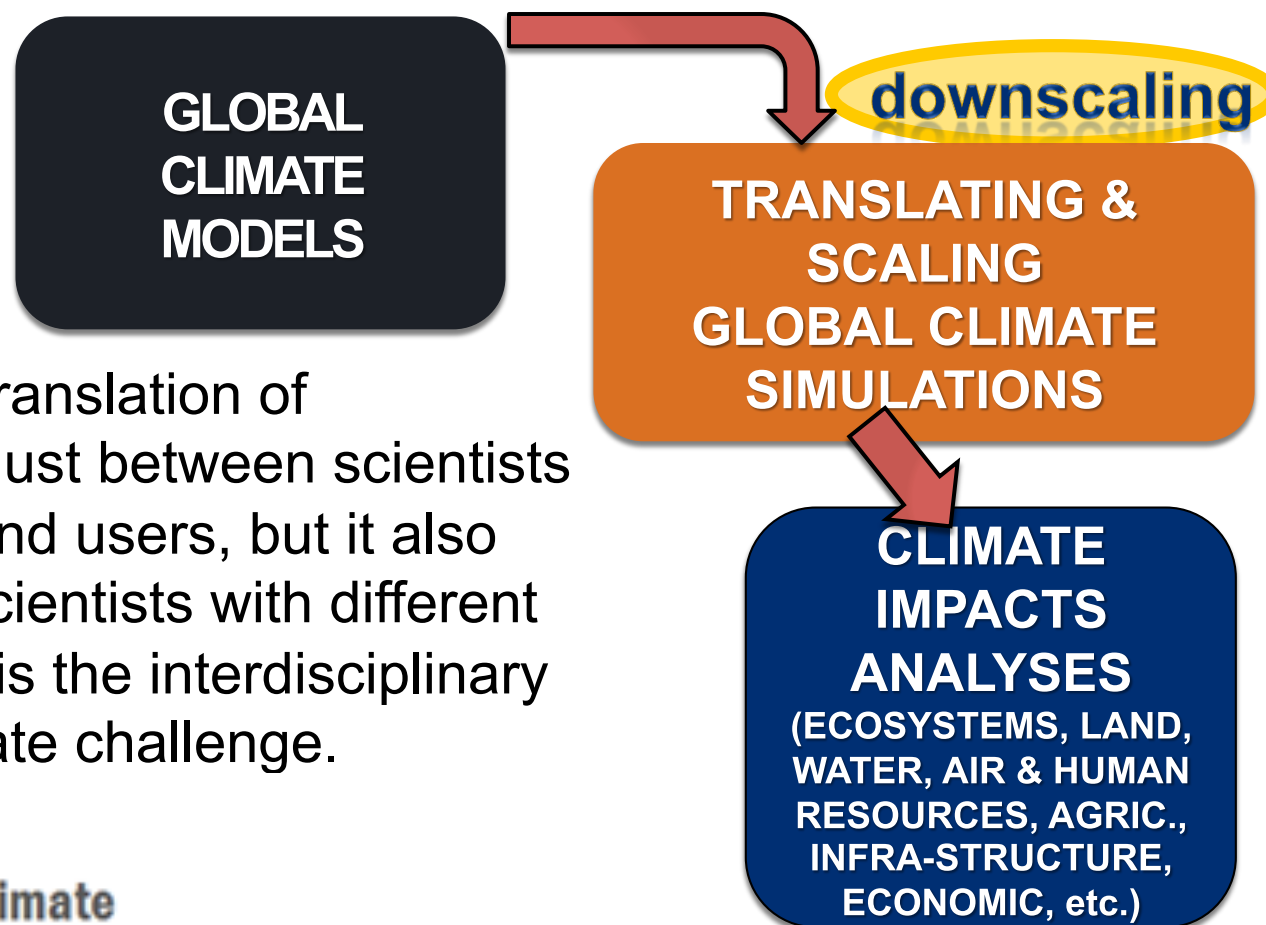
# FROM SCIENCE TO STAKEHOLDERS

## The Transfer & Translation of Climate-Relevant Information



# FROM SCIENCE TO STAKEHOLDERS

## The Transfer & Translation of Climate-Relevant Information

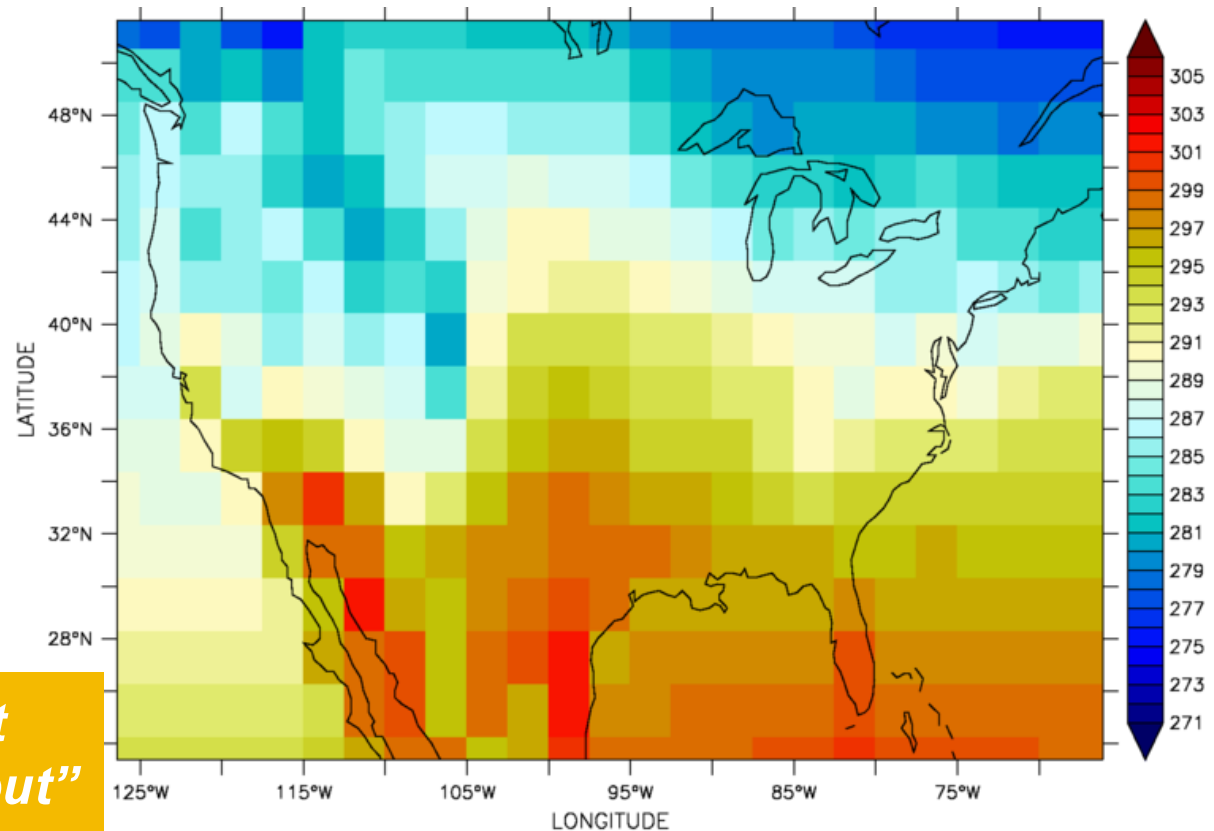


The transfer and translation of information is not just between scientists and the ultimate end users, but it also occurs between scientists with different specialties - such is the interdisciplinary nature of the climate challenge.

# Why do statistical downscaling?

(aka **ESD**:  
empirical  
statistical  
downscaling)

*“statistical refinement  
of dynamical model output”*



Annual Mean Daily Maximum Temperature [K] (01 JAN 1979 - 31 DEC 2008)

## Common expectations for statistically downscale climate projections:

- (a) Statistical downscaling adds realistic, finer-scale detail not found in the coarser resolution global climate models (GCMs).
- (b) Statistical downscaling aims to correct for GCM biases and other shortcomings in the simulated distribution of climate variables of interest.

# Why do statistical downscaling?

(aka **ESD**:  
empirical  
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## ESD: Somewhat analogous to the MOS

Instead of maintaining a dynamic climate model at higher resolution, the statistical downscaling approach refines information from GCMs by using a series of empirically-derived equations to relate variations in large-scale climate to variations in local climate.

Adapted from  
[www.southwestclimatechange.org/climate/modeling/downscaling](http://www.southwestclimatechange.org/climate/modeling/downscaling)

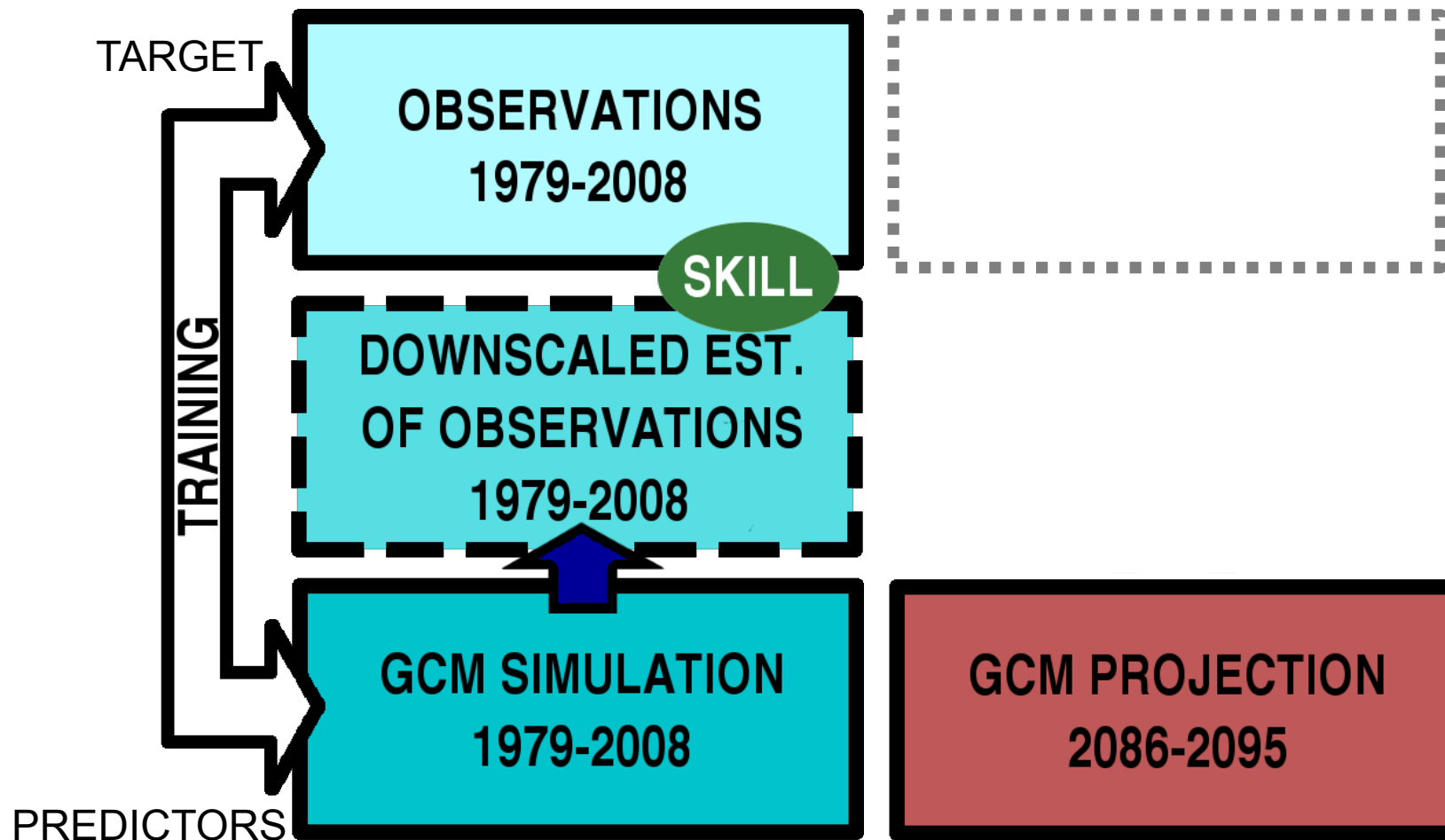
***Assumed to be “value-added” product***

## Common expectations for statistically downscale climate projections:

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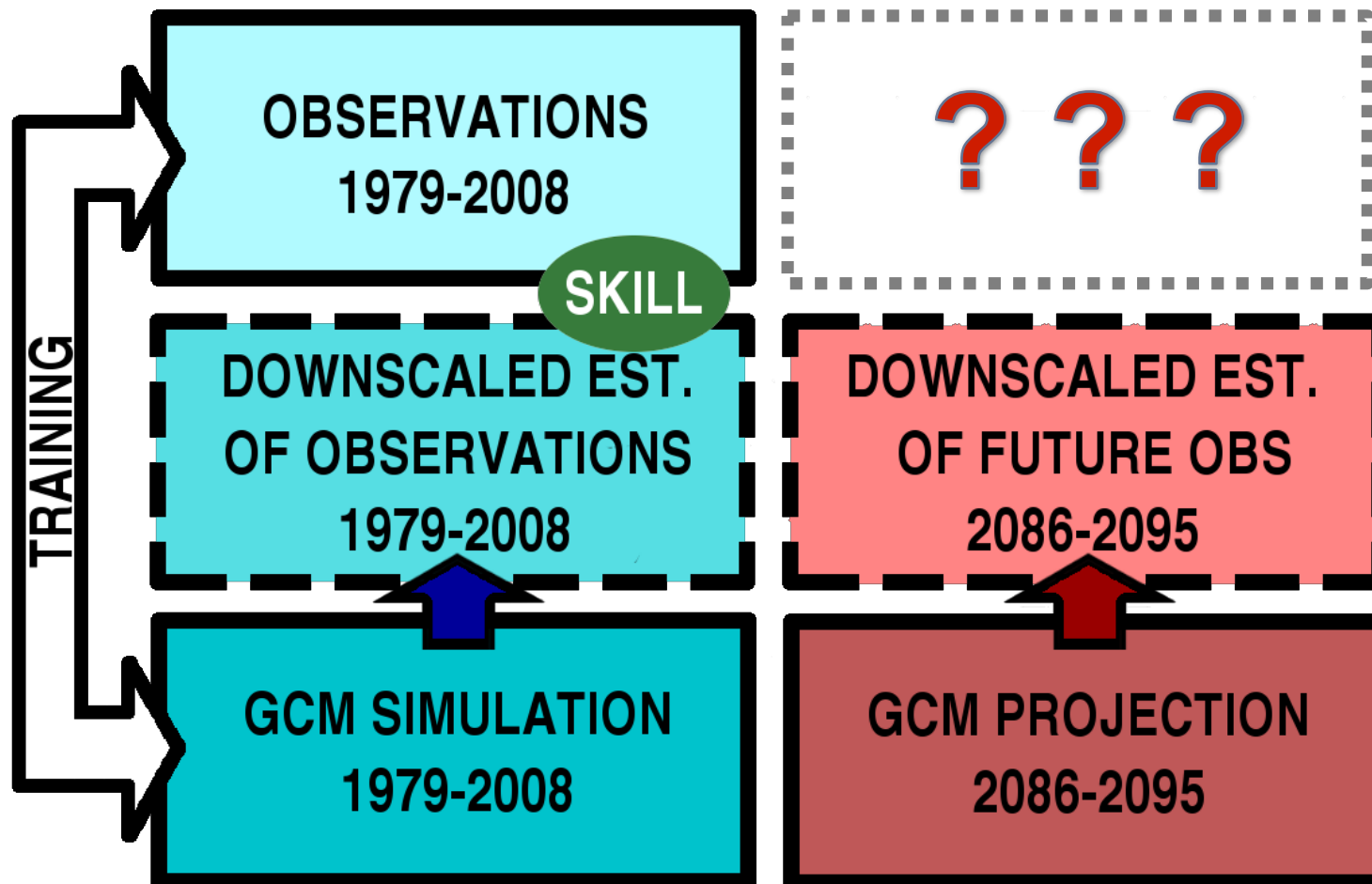
# REAL WORLD APPLICATION:

**Skill: Compare Obs to SD historical output (e.g., cross validation)**



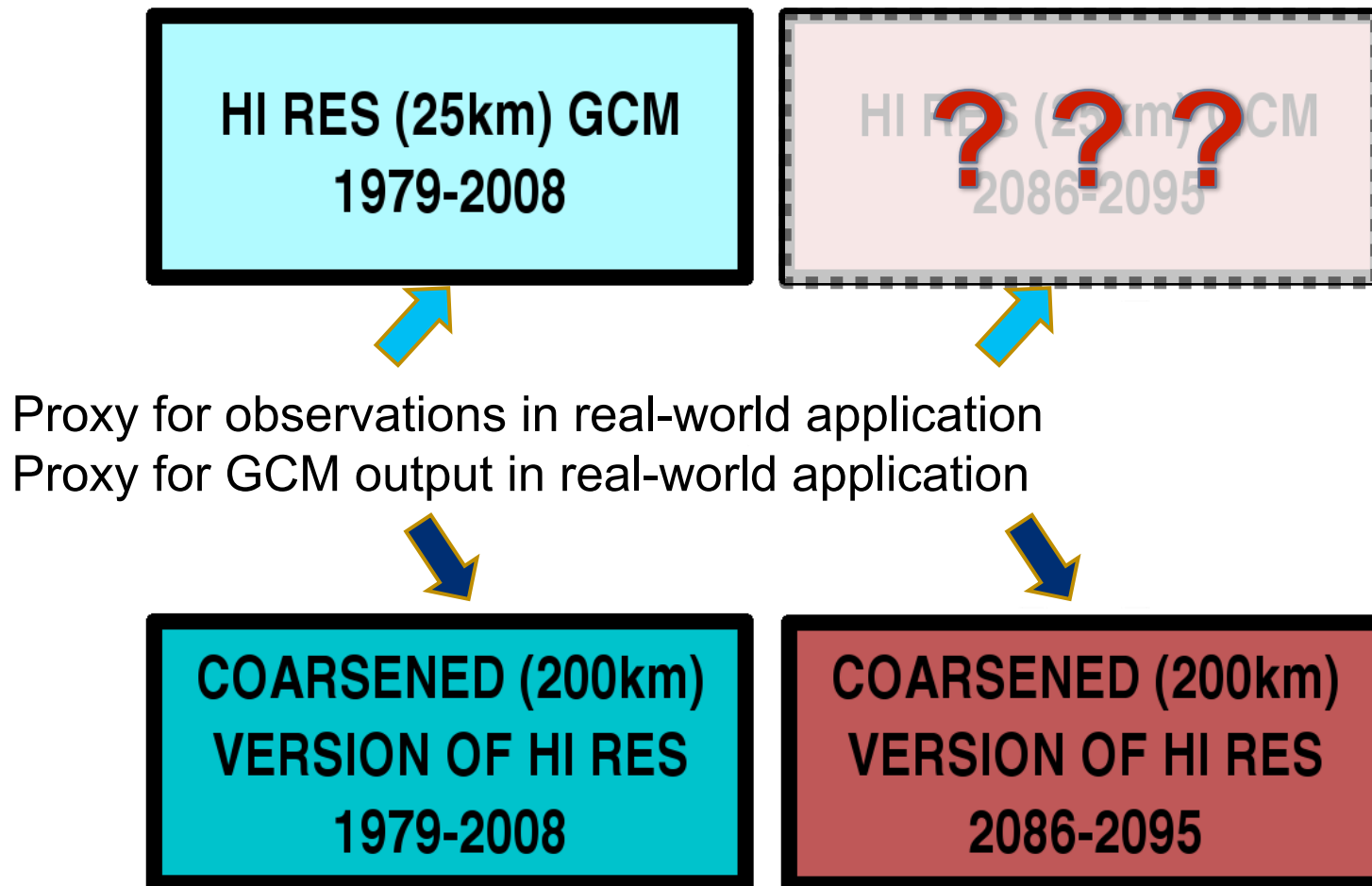
# REAL WORLD APPLICATION:

Cannot evaluate future skill -- left assuming transform functions apply equally well to past & future -- **“The Stationarity Assumption”**



# “PERFECT MODEL” EXPERIMENTAL DESIGN:

Start with 4 types of data sets – Hi-res GCM output as proxy for obs & coarsened version of Hi-res GCM output as proxy for usual GCM





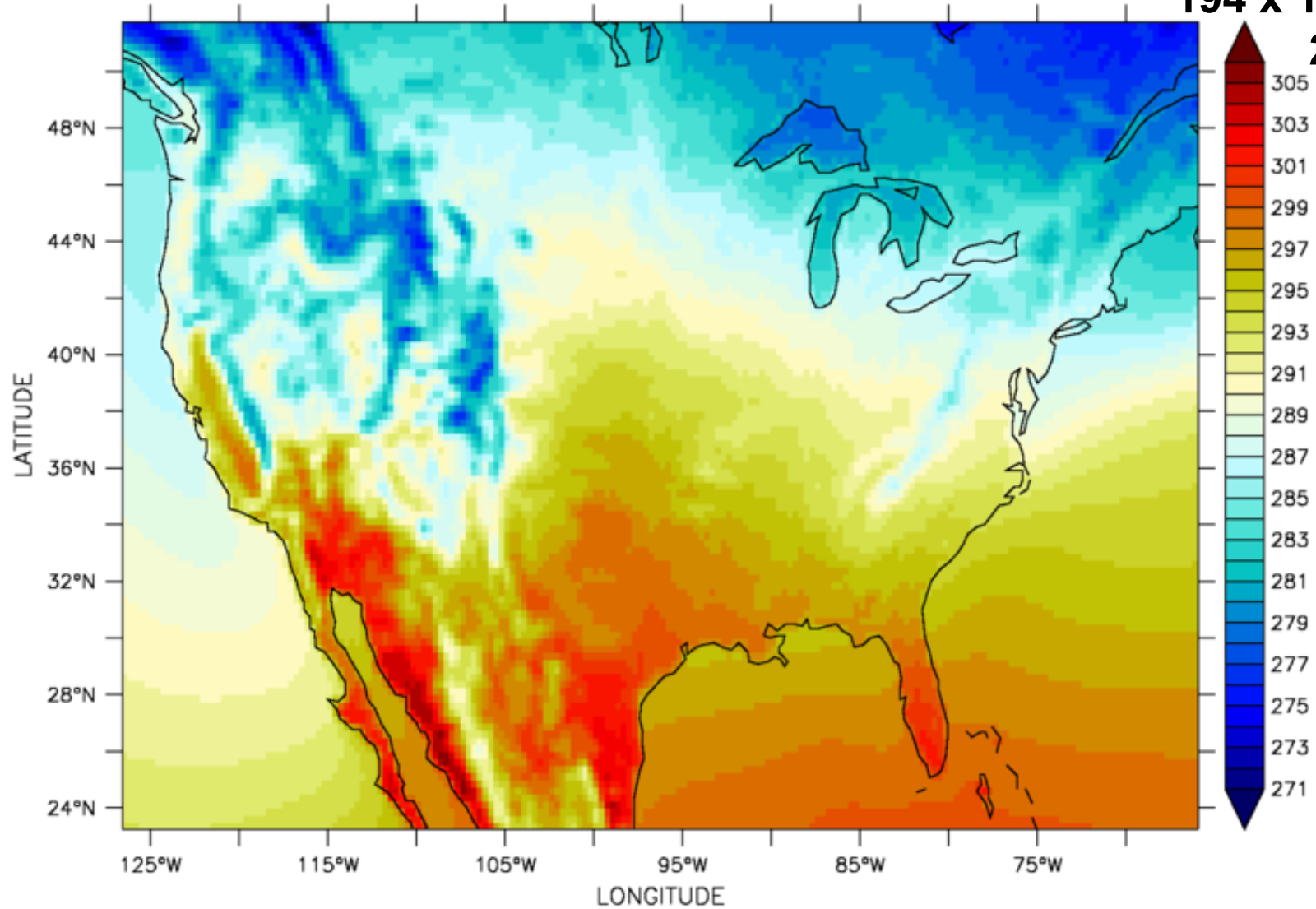
**(a) GFDL-HIRAM-C360 hi-res model (psuedo-obs)**

**Daily time resolution**

**~25km grid spacing**

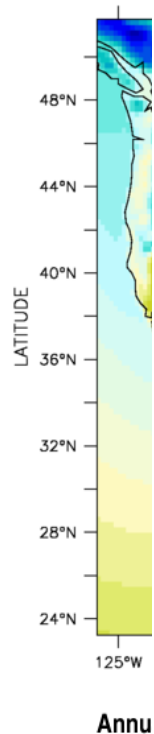
**194 x 114 grid**

**22k pts**



**Annual Mean Daily Maximum Temperature [K] (01 JAN 1979 - 31 DEC 2008)**

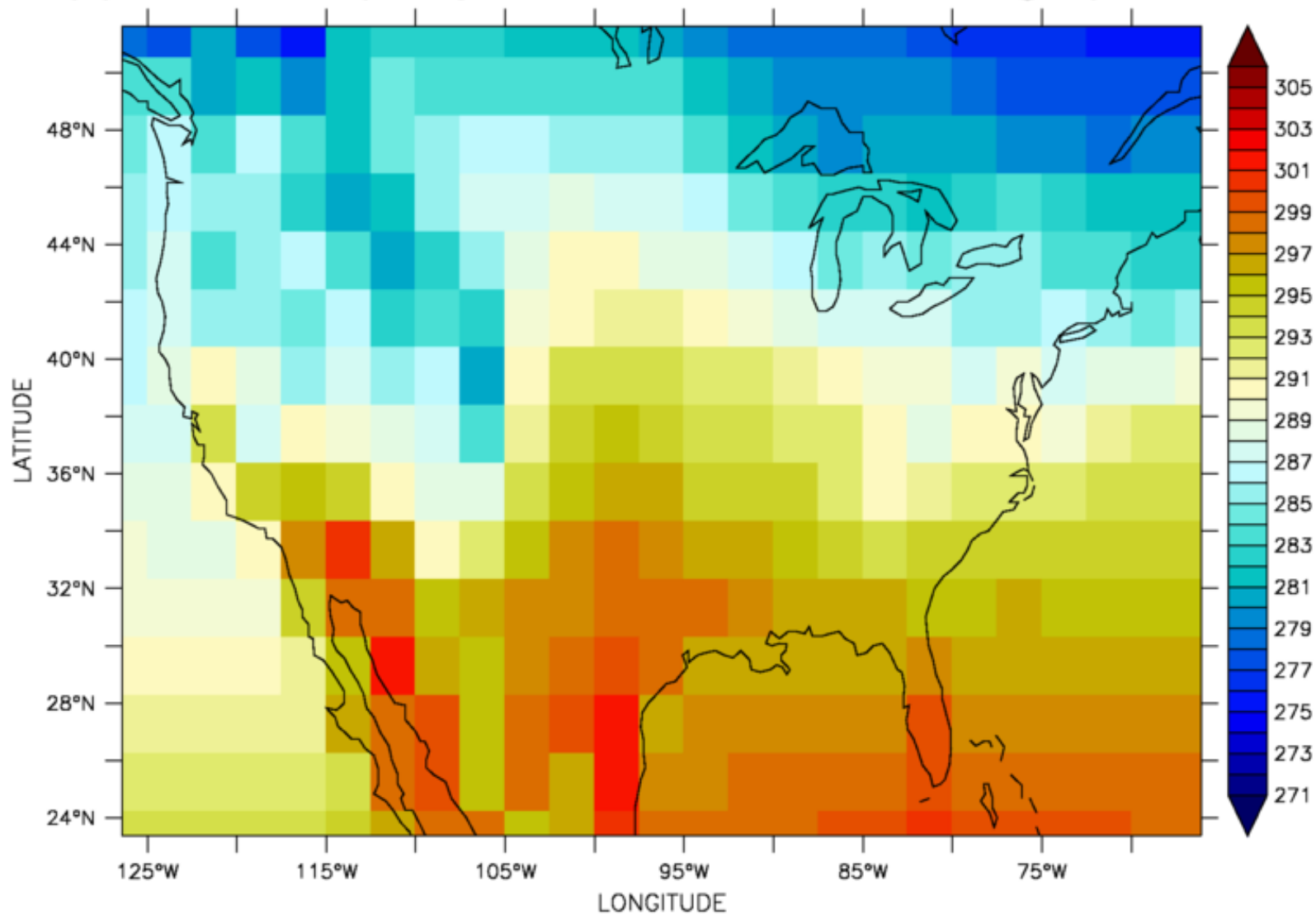
(a) GF



64:1

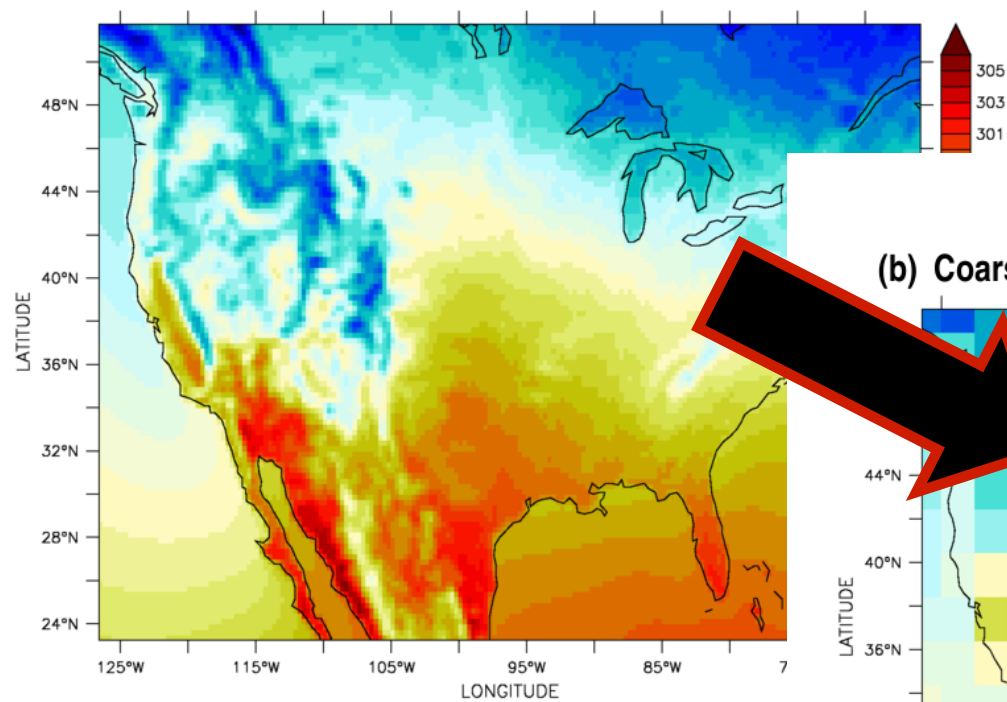
~200km grid spacing

(b) Coarsened (interpolated from hi-res to coarse grid)



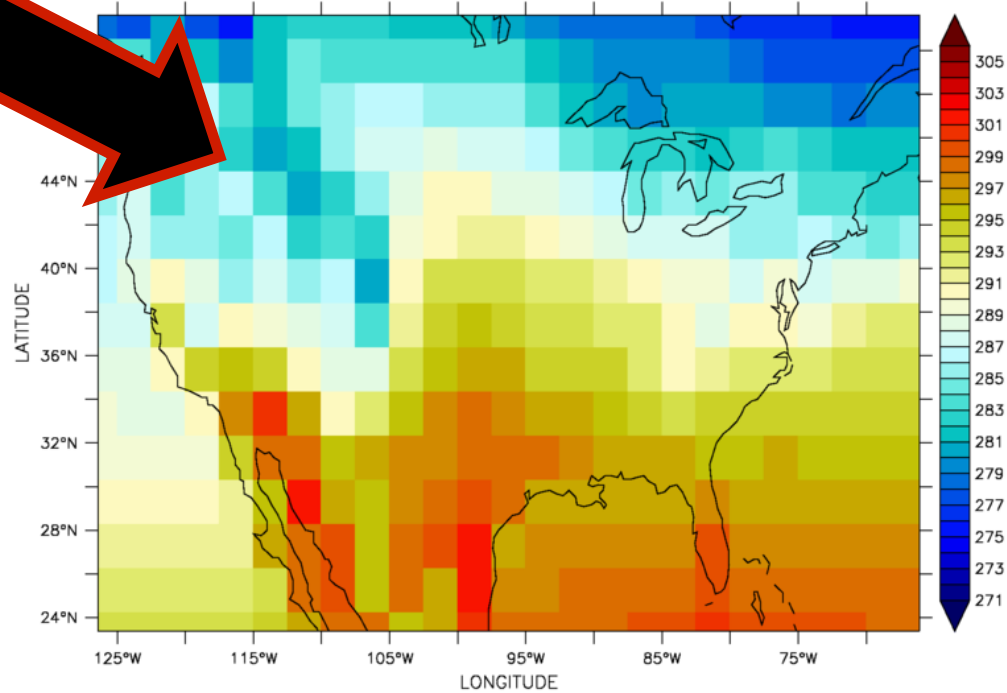
Annual Mean Daily Maximum Temperature [K] (01 JAN 1979 - 31 DEC 2008)

**(a) GFDL-HIRAM-C360 hi-res model (psuedo-obs)**



**Annual Mean Daily Maximum Temperature [K] (01 JAN 1979 - 31 DEC 2008)**

**(b) Coarsened (interpolated from hi-res to coarse grid)**



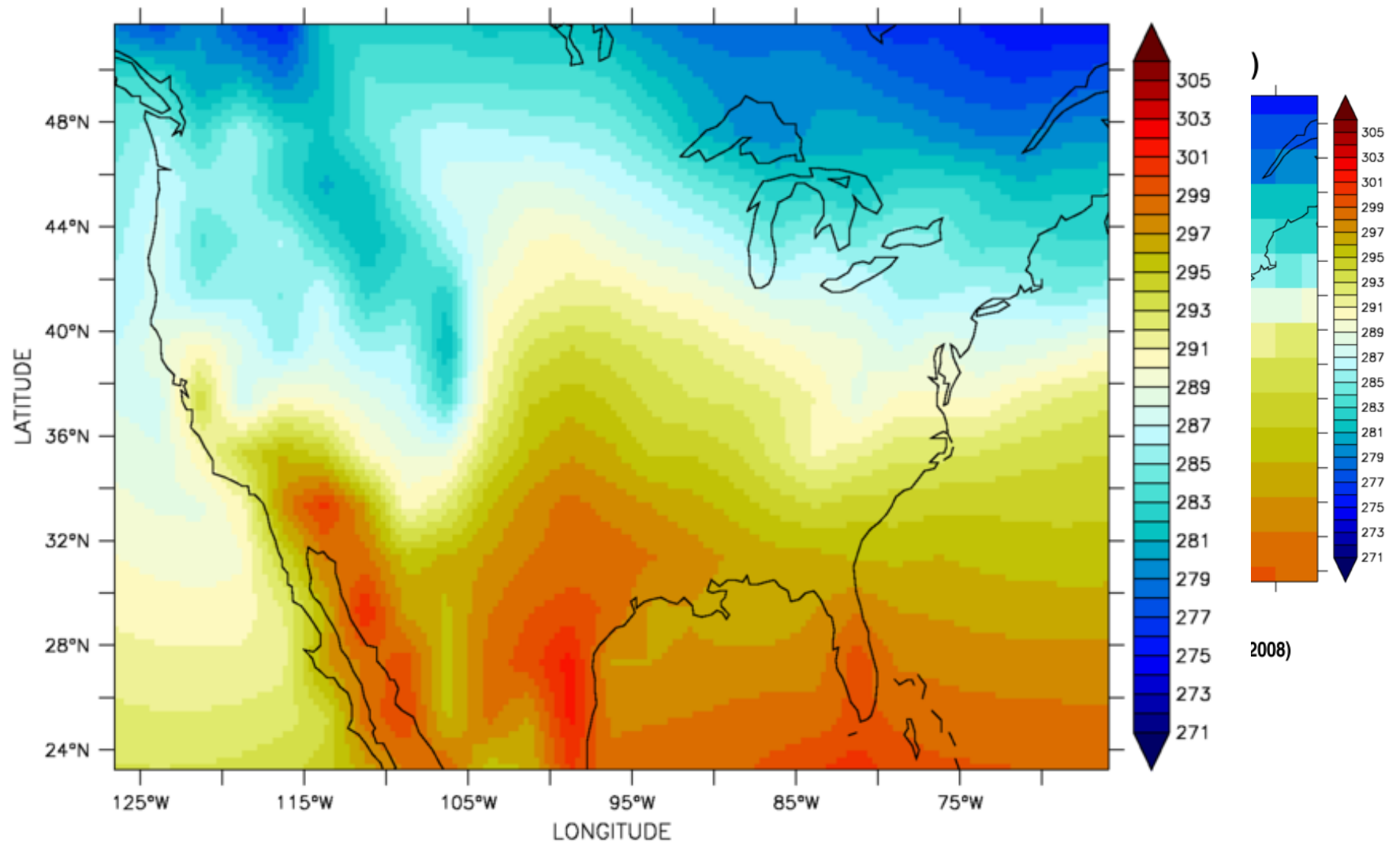
**Annual Mean Daily Maximum Temperature [K] (01 JAN 1979 - 31 DEC 2008)**

(a) GFDL-HIRAM-C360 hi-res model (pseudo-obs)

Daily time resolution  
~25km grid spacing

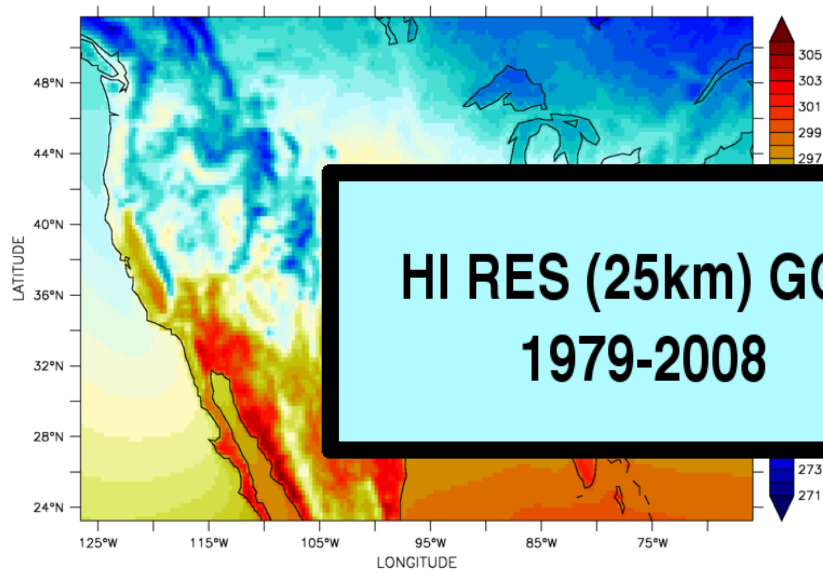
(c) Regrided (interpolated from coarse to hi-res grid)

194 x 114 grid



Annual Mean Daily Maximum Temperature [K] (01 JAN 1979 - 31 DEC 2008)

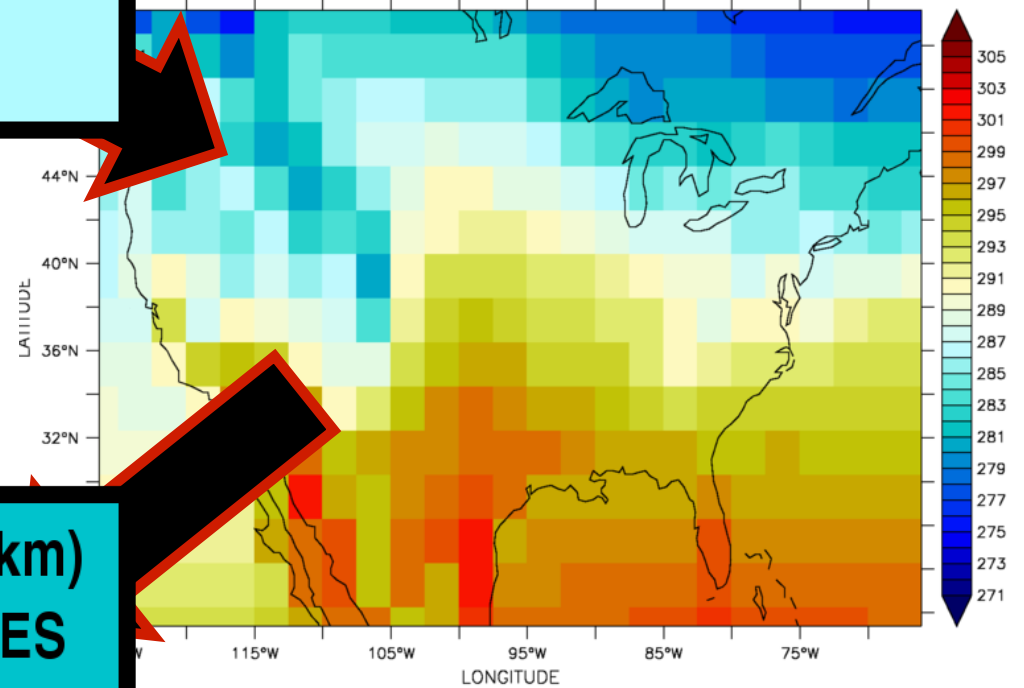
(a) GFDL-HIRAM-C360 hi-res model (psuedo-obs)



Annual Mean Daily Maximum Temperature [K] (01 JAN 1979 - 31 DEC 2008)

**HI RES (25km) GCM  
1979-2008**

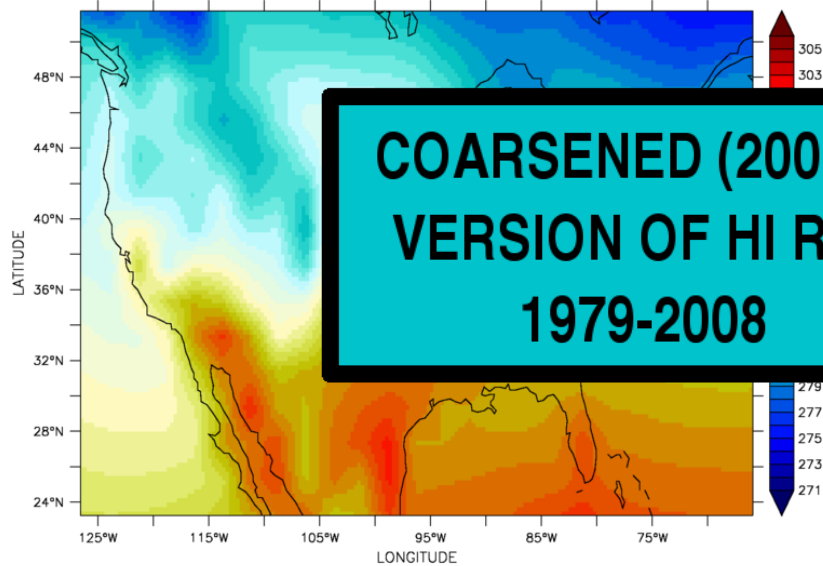
Coarsened (interpolated from hi-res to coarse grid)



Annual Mean Daily Maximum Temperature [K] (01 JAN 1979 - 31 DEC 2008)

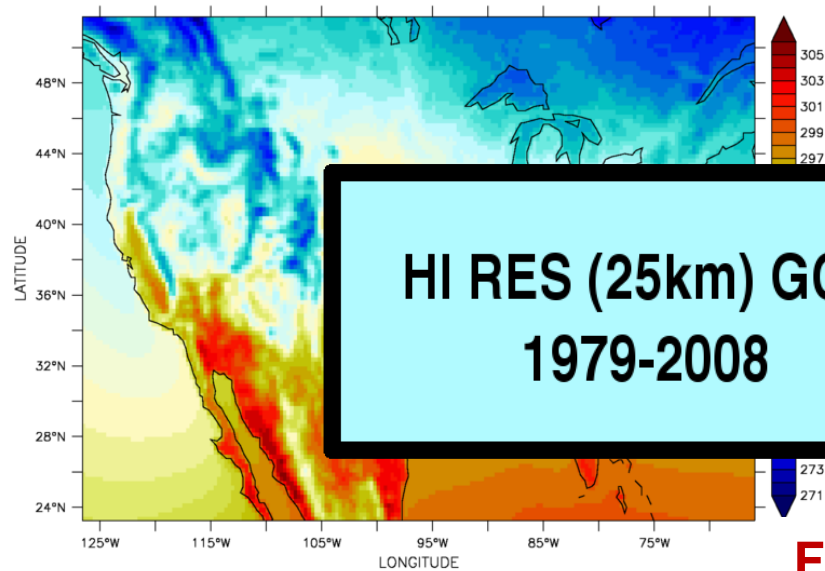
**COARSENEDED (200km)  
VERSION OF HI RES  
1979-2008**

(c) Regridded (interpolated from coarse to hi-res grid)

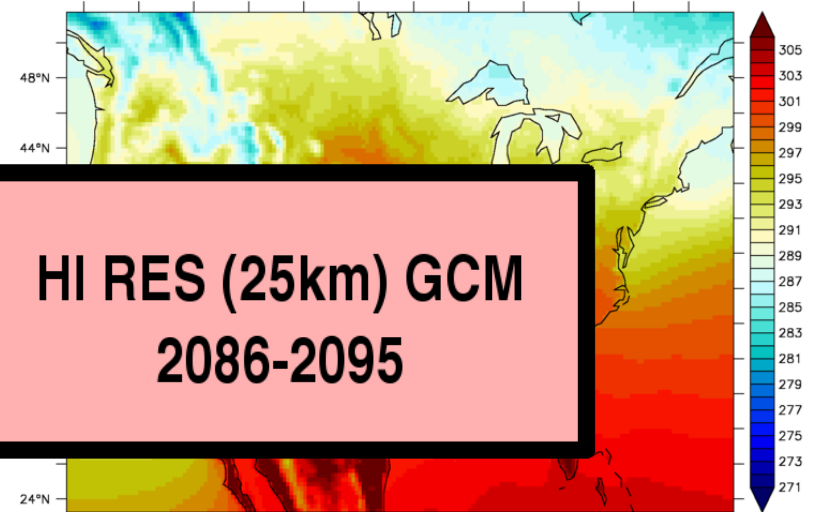


Annual Mean Daily Maximum Temperature [K] (01 JAN 1979 - 31 DEC 2008)

(a) GFDL-HIRAM-C360 hi-res model (psuedo-obs)



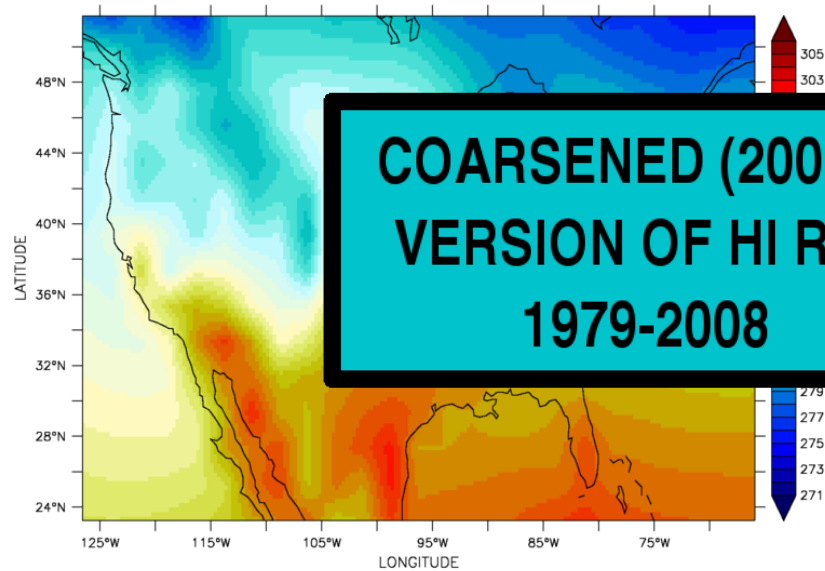
**HI RES (25km) GCM  
1979-2008**



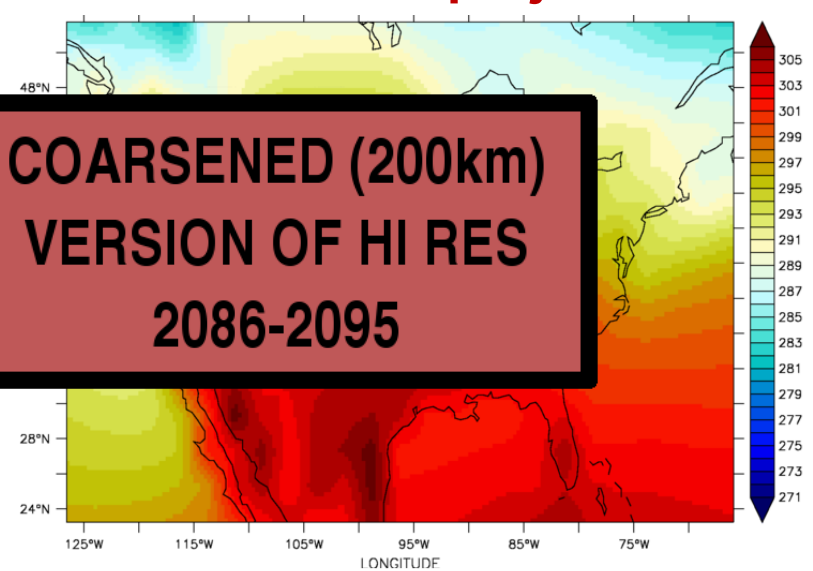
**HI RES (25km) GCM  
2086-2095**

**Follow the same interpolation/regridding  
sequence to produce coarsened data  
sets for the future climate projections**

(c) Regrided (interpolated from coarse to hi-res grid)



**COARSENEED (200km)  
VERSION OF HI RES  
1979-2008**

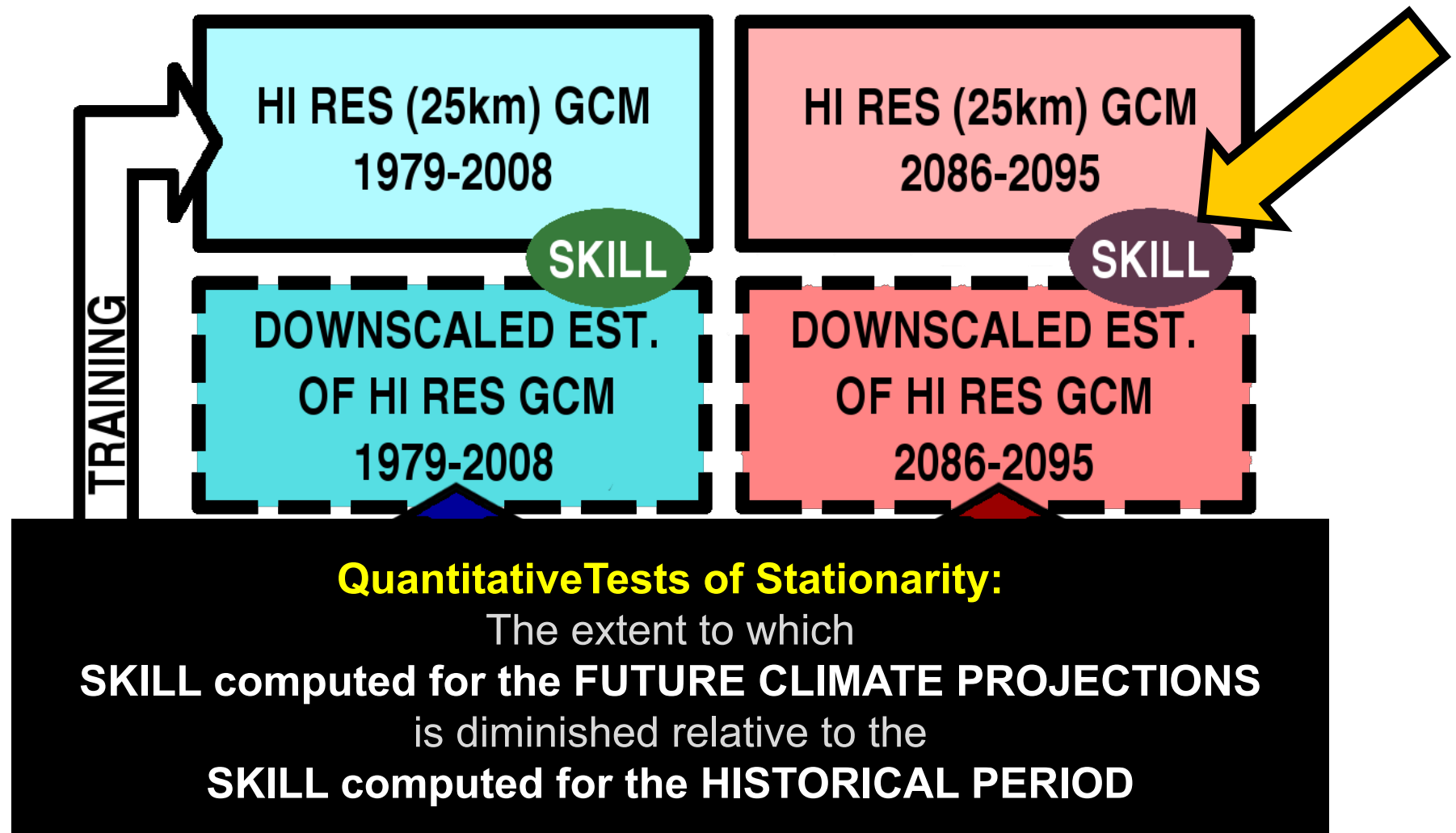


**COARSENEED (200km)  
VERSION OF HI RES  
2086-2095**

Annual Mean Daily Maximum Temperature [K] (01 JAN 1979 - 31 DEC 2008)

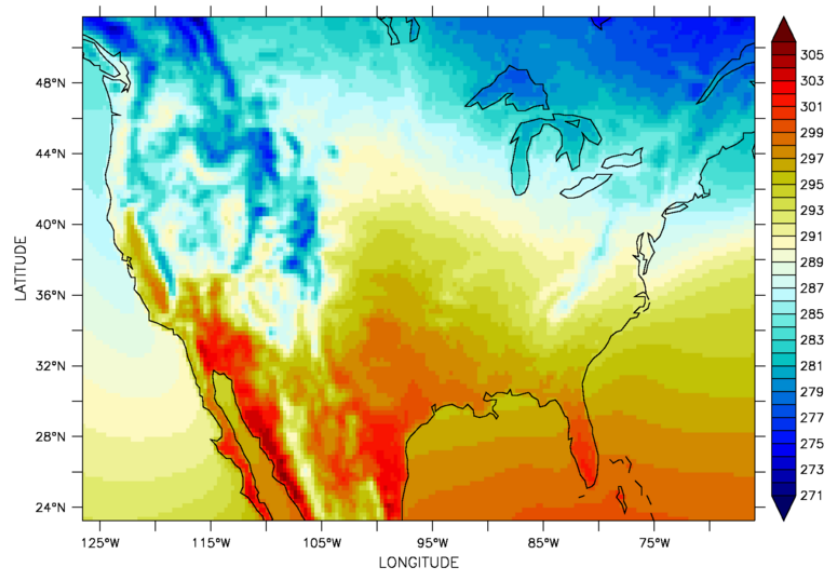
# “PERFECT MODEL” EXPERIMENTAL DESIGN:

Can directly evaluate skill both for the historical period and the future using the Hi Res GCM output as “truth” – **Test Stationarity**



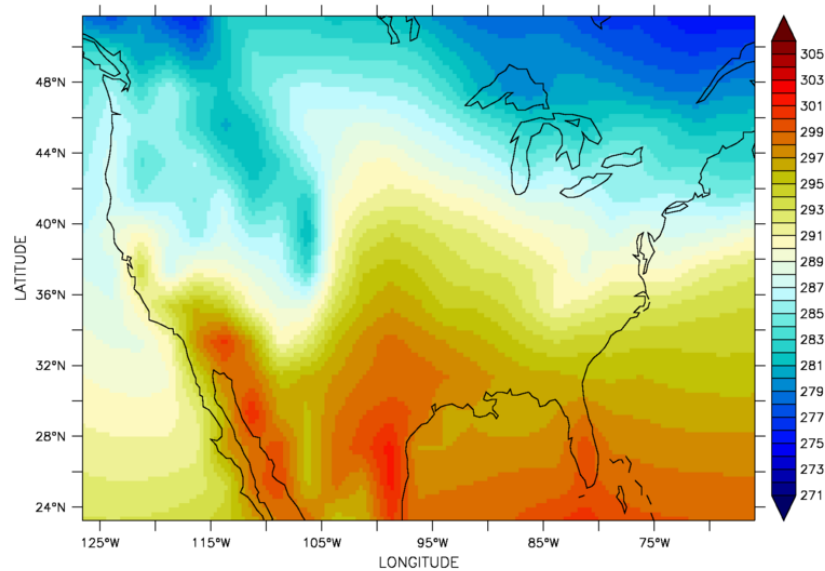


(a) GFDL-HIRAM-C360 hi-res model (psuedo-obs)



Annual Mean Daily Maximum Temperature [K] (01 JAN 1979 - 31 DEC 2008)

(c) RegridDED (interpolated from coarse to hi-res grid)



Annual Mean Daily Maximum Temperature [K] (01 JAN 1979 - 31 DEC 2008)

**Q:** How High a Hurdle does this Perfect Model approach present?

**A:** Varies geographically, by variable of interest, time period of interest, size of GCM-projected climate response, etc.

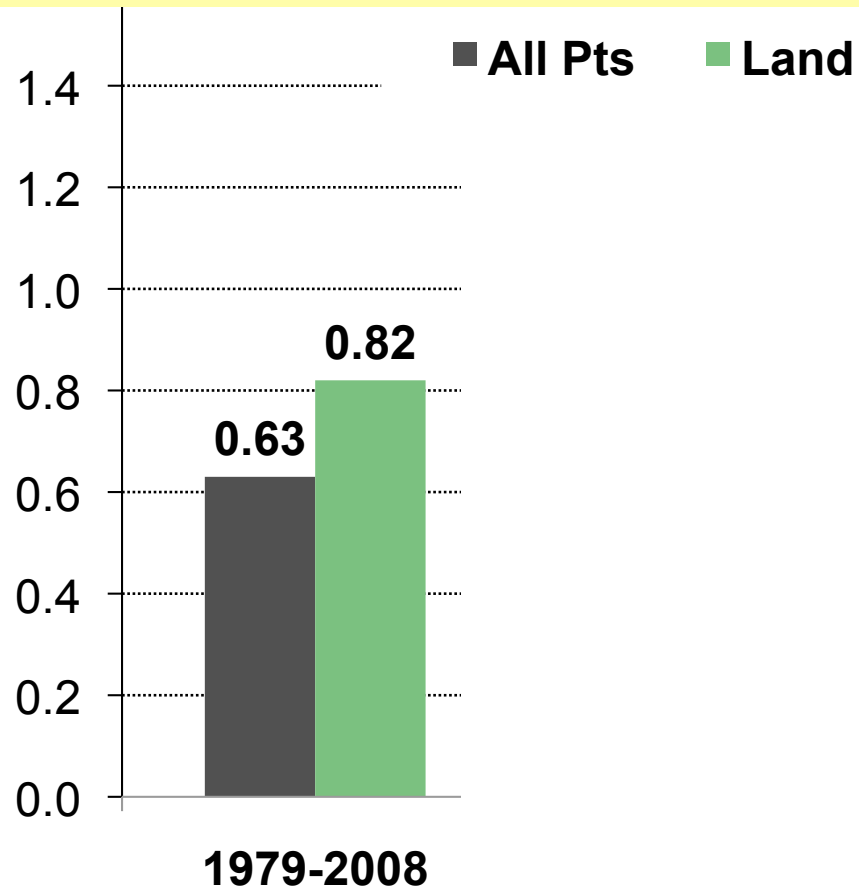
**And it will vary among ESD techniques**



## **NEXT:** A sampling of results...

- Intended to be illustrative  
Not exhaustive, nor systematic.

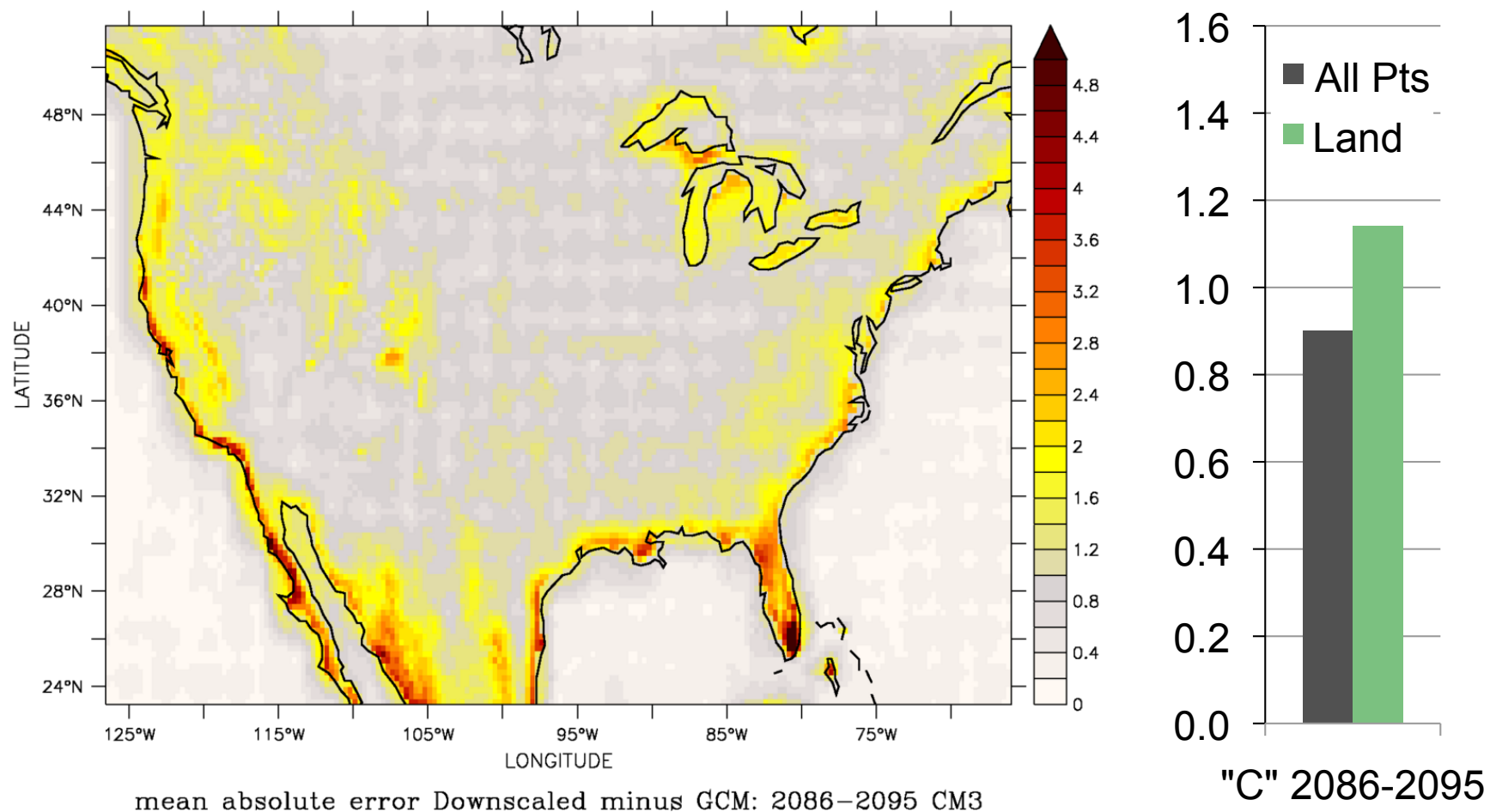
**Start with a summary: ARRM downscaling errors are larger for daily max temp at end of 21<sup>st</sup>C than for 1979-2008**



**Area Mean Time Mean Absolute Downscaling Errors**

$$\frac{\sum |(\text{Downscaled Estimate} - \text{HiRes GCM})|}{(\text{NumDays})}$$

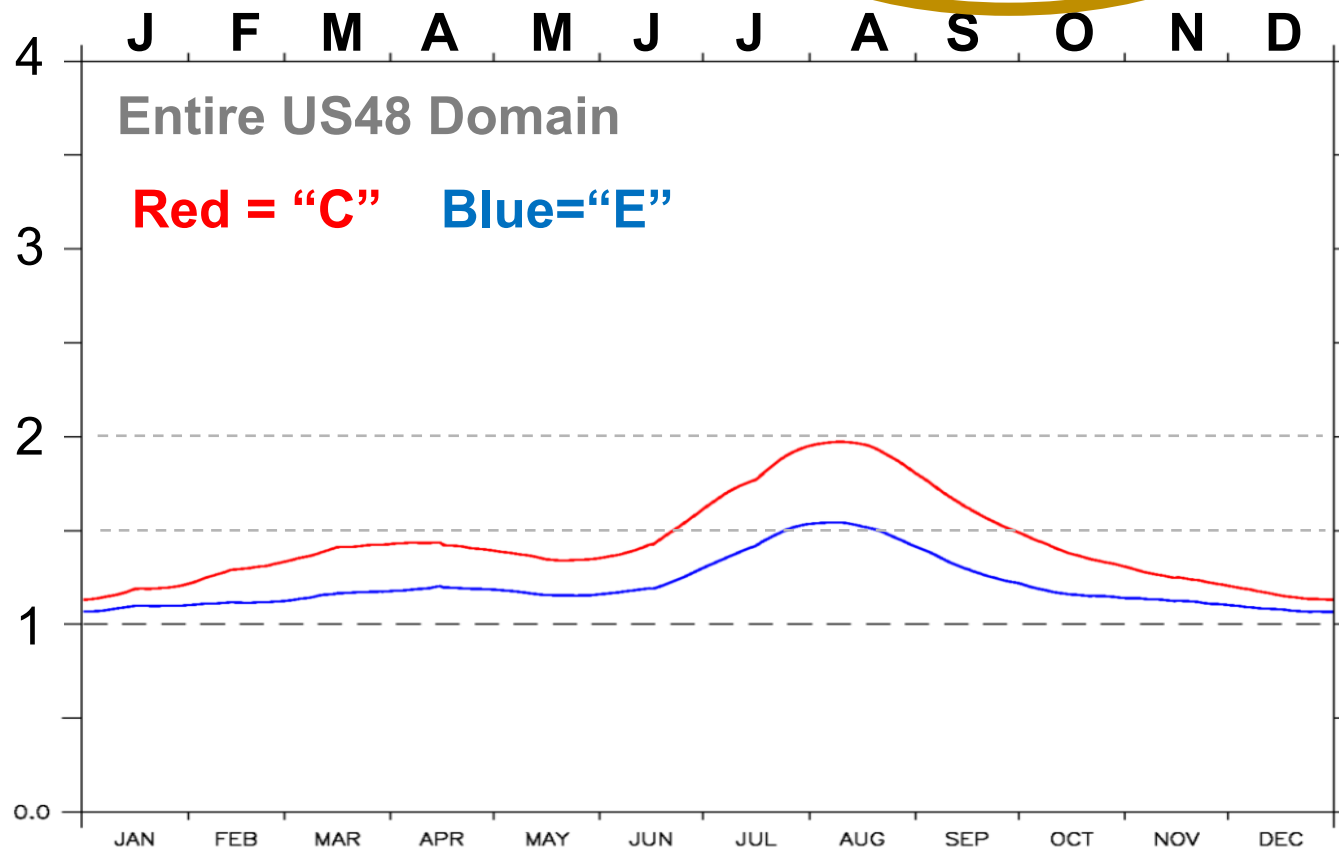
## Geographic Variations: MAE pattern for “C” projections (+7,6C)



**Mean absolute downscaling error during 2086-2095**  
“C” Projections (3 member ensemble)

Looking at how well the stationarity assumption holds in different seasons

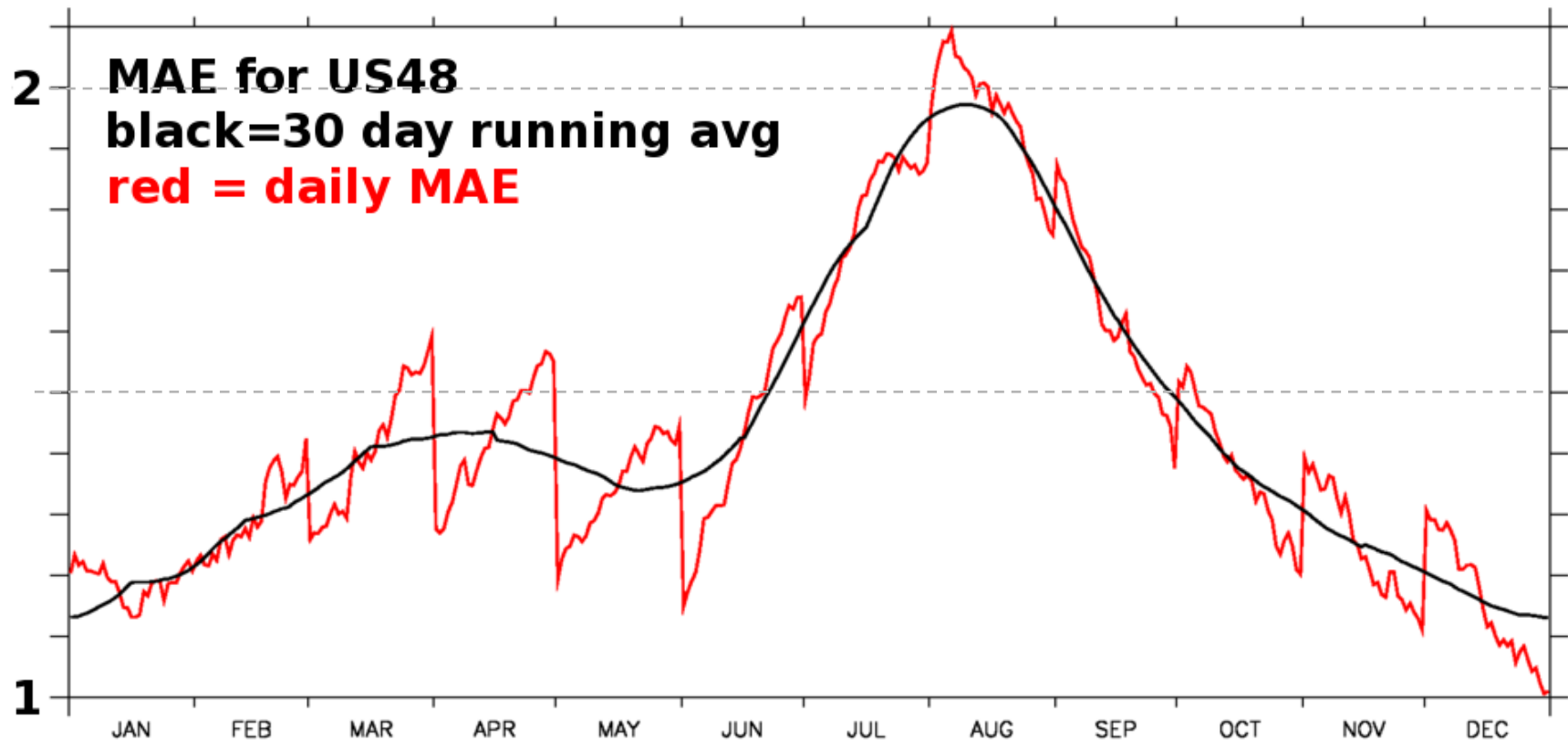
ARRM method  
31-day running mean MAE ratios  
tasmax = daily max temp @ 2m



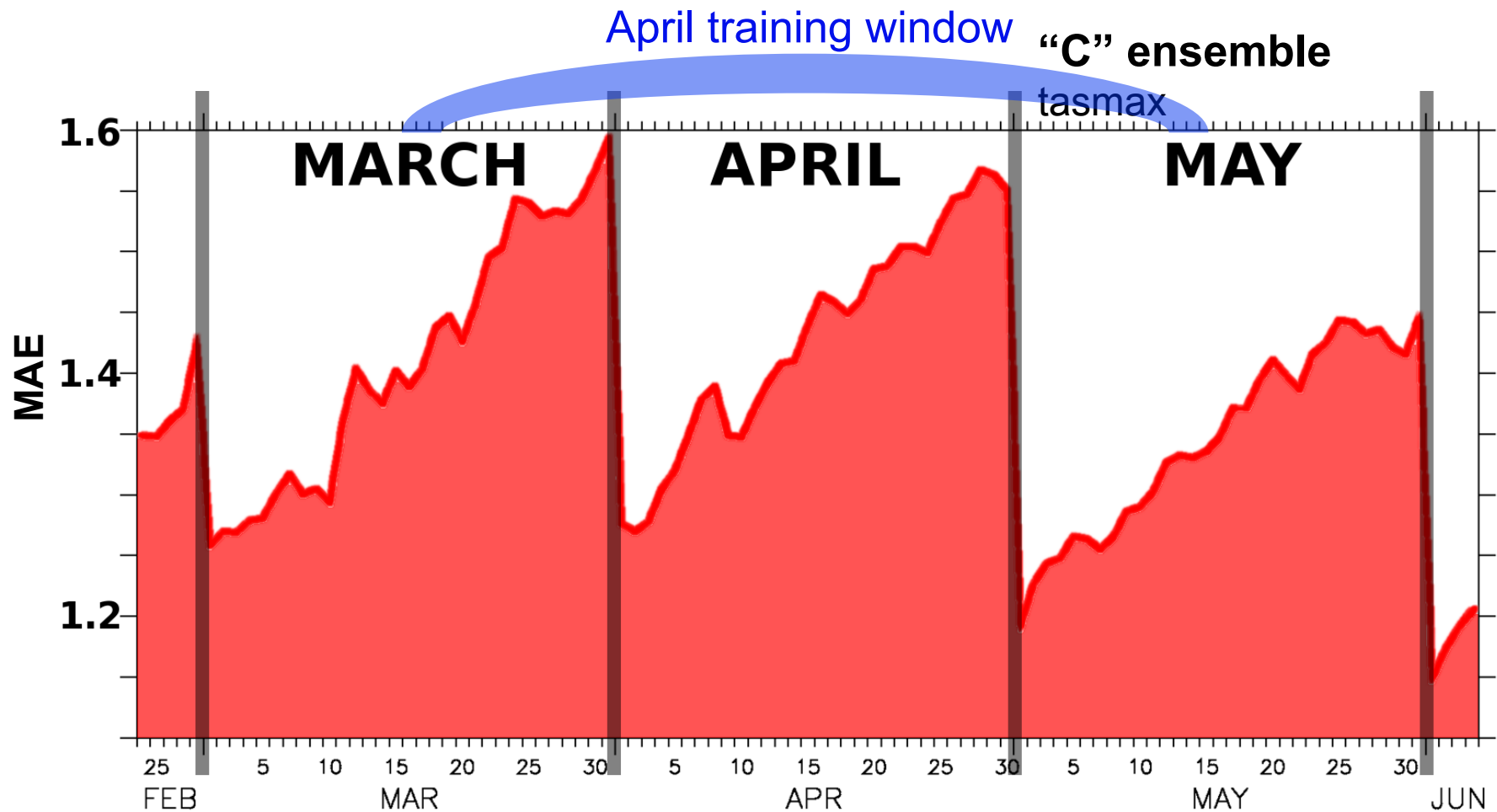
**If and when ratio=1.0, the stationarity assumption fully holds**  
(i.e., no degradation in mean absolute downscaling error  
during 2085-2095 vs. the 1979-2008 period using in training.)

Looking at how well the stationarity assumption holds in different seasons

“C” ensemble  
tasmax

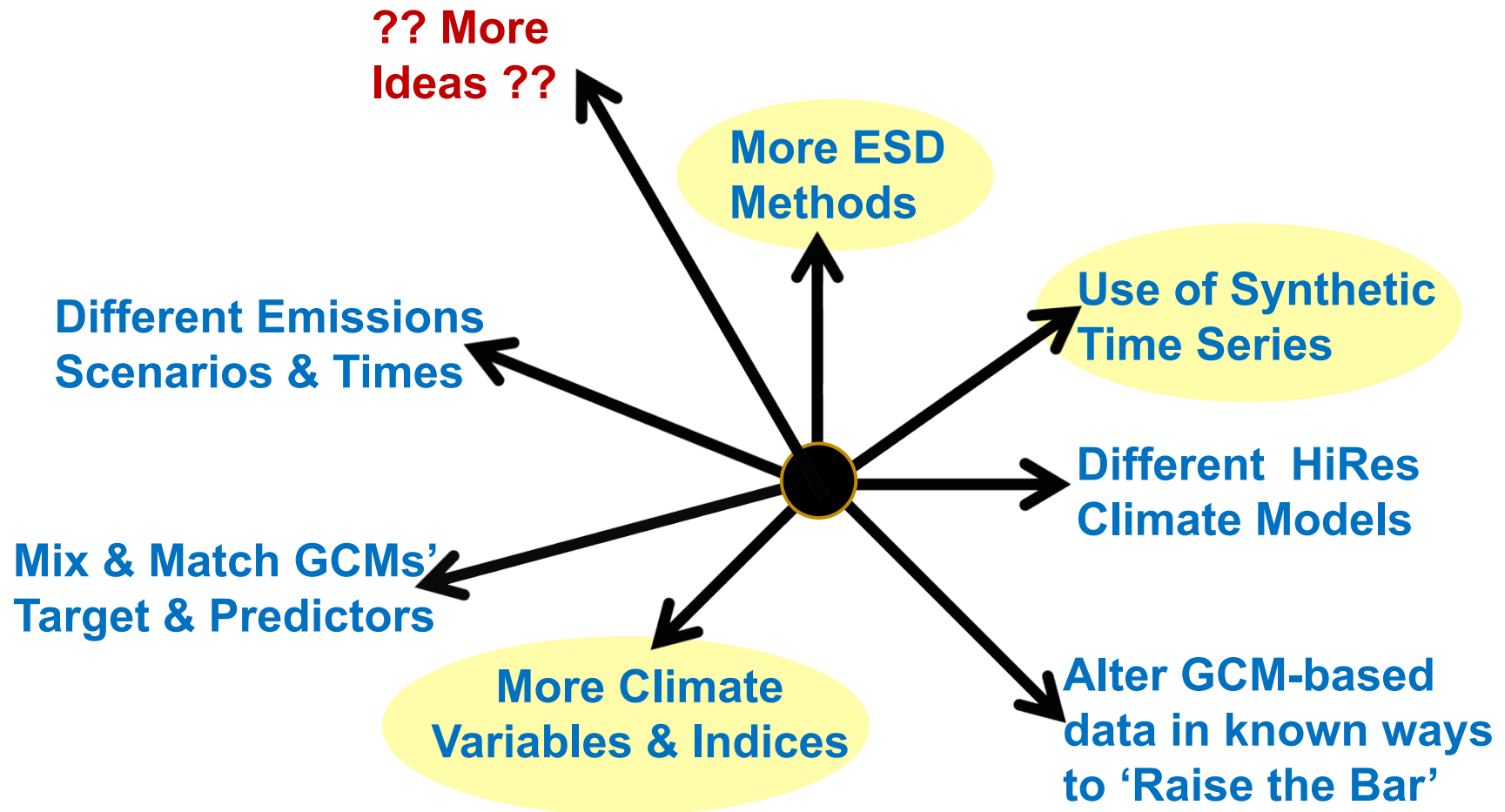


A clear intra-month MAE trend in some but not all months



“sawtooth” has smaller errors in the cooler part of the month –and- larger errors in the warmer part of month, when applied to end of 21<sup>st</sup> century projections

## Options for extending this 'perfect model-based' exploration of statistical downscaling stationarity



# Goals of this presentation

1. Define the 'stationarity assumption' inherent to statistical downscaling future climate projections.
2. Present our 'perfect model' (aka 'big brother') approach to quantitatively assess the extent to which the stationarity assumption holds.
3. Illustrate with a few examples the kind of results one can generate using this evaluation framework
4. Introduce approaches to extend and supplement the method (setting the hurdle at different heights).
5. Invite statistical downscalers to consider testing their methods within the perfect model framework.

[www.gfdl.noaa.gov/esd](http://www.gfdl.noaa.gov/esd)





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